

Chapter 5

Hypotheses Test of the Mean

Section 5.2

Conduct the Hypothesis Test

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- Conduct the Hypothesis Test
 - Compare Obtained t_m to $t_{\alpha/2}$
 - Compare Obtained p -value to α
 - Application
 - Appendix: Obtain the p -value Directly

5.2a

Compare Obtained t_m to $t_{\alpha/2}$ Criterion

Illustration: t in the Rejection Region

Calculate the t -statistic

- ▶ Suppose a claim, the **null hypothesis**, is that some variable Y has a population mean of 5, $H_0 : \mu = 5$
- ▶ Consider an actual **sample of data** of variable Y for $n = 60$ that yields calculated values of $m = 5.76$ and $s = 2.73$
- ▶ Is $m = 5.76$ **so much larger than the reference value** of 5 that $\mu = 5$ is unreasonable?
- ▶ To answer, first estimate the **standard error of the mean**

$$s_m = \frac{s}{\sqrt{n}} = \frac{2.73}{\sqrt{60}} = 0.352$$

- ▶ Then obtain the observed **standardized distance of m from μ_0**

$$t_m = \frac{m - \mu_0}{s_m} = \frac{5.76 - 5}{0.352} = 2.16$$

Illustration: t in the Rejection Region

- ▶ **Criterion:** Start with the **theoretical t -distribution** and $\alpha = 0.05$ criterion without any reference to the data
- ▶ **From the data:** $t_m = 2.16$, so the sample mean is 2.16 **estimated standard errors** above the hypothesized mean
- ▶ **Evaluate:** compare obtained $t_m = 2.16$ with **upper tail cutoff** $t_{.025; df=59} = 2.00$

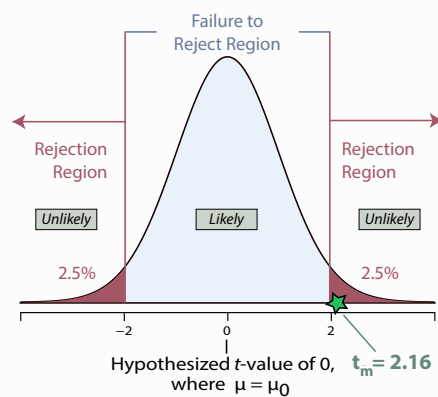


Illustration: t in the Rejection Region

- ▶ For $df = 59$, $t_m = 2.16 > t_{.025} = 2.00$, so t_m lies in the **upper tail rejection region**
- ▶ IF $H_0 : \mu = \mu_0$ is true, THEN obtained outcome of t_m is unlikely, so **reject $H_0 : \mu = 5$**
- ▶ Conclude that $\mu = 5$ is **unreasonable**

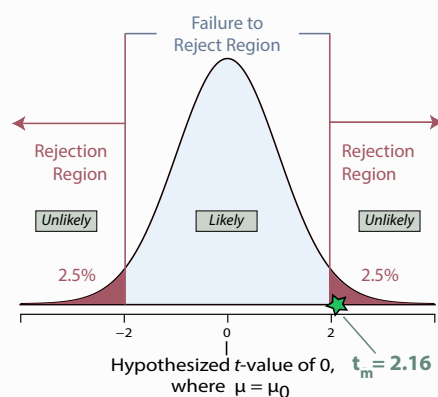


Illustration: t NOT in the Rejection Region

Calculate the t -statistic

- Suppose the claim is the null hypothesis for some variable Y is that its population mean is 5, $H_0 : \mu = 5$
- Consider an actual sample of data of variable Y for $n = 60$ that yields calculated values of $m = 5.37$ and $s = 2.73$
- Is $m = 5.37$ so much larger than the reference value of 5 that $\mu = 5$ is not reasonable?
- First estimate the standard error of the mean

$$s_m = \frac{s}{\sqrt{n}} = \frac{2.73}{\sqrt{60}} = 0.352$$

- Then obtain the observed standardized distance of m from μ_0

$$t_m = \frac{m - \mu_0}{s_m} = \frac{5.37 - 5}{0.352} = 1.05$$

Illustration: t NOT in the Rejection Region

- **Criterion:** Start with the theoretical t -distribution and $\alpha = 0.05$ criterion without any reference to the data
- **From the data:** $t_m = 1.05$, so the sample mean is 1.05 estimated standard errors above the hypothesized mean
- **Evaluate:** compare obtained $t_m = 1.05$ with upper tail cutoff $t_{.025; df=59} = 2.00$

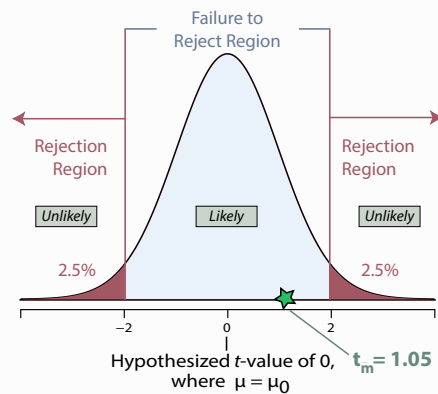
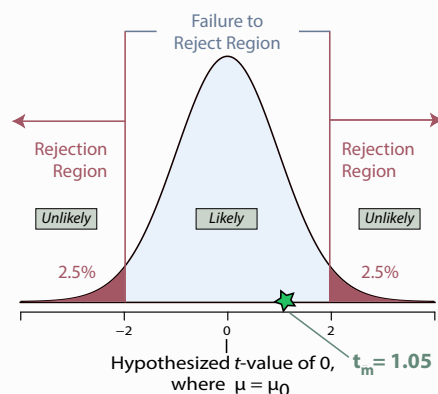


Illustration: t NOT in the Rejection Region

- For $df = 59$, $t_m = 1.05 < t_{.025} = 2.00$, so t_m does NOT lie in a rejection region
- IF $H_0 : \mu = \mu_0$ is true, THEN t_m is reasonably likely, so do not reject $H_0 : \mu = 5$
- Conclude: No difference from $\mu = 5$ detected
- Do not conclude that $\mu = 5$ is the correct value



5.2b

Compare Obtained p -value to α Criterion

Assess How Close m is to μ_0 with Probability

IF μ_0 is true, THEN how probable is the sample value t_m ?

- ▶ **p -value:** Given a true null hypothesis, the probability of obtaining a value as far from or farther from the hypothesized value the value of the obtained sample mean
- ▶ Usually either a large positive or a large negative deviation from the hypothesized value is of interest, so the p -value usually assesses the probability of a value in either tail
- ▶ IF $\mu = \mu_0$, THEN
 - A low p -value indicates an unlikely sample mean, so t_m is in the rejection region, which means m is far from μ_0
 - A high p -value indicates a likely sample mean, so t_m is not in the rejection region, and m is reasonably close to μ_0
- ▶ Prefer more convenient probability expression of the p -value because probabilities are more generally understood than a t -value, and both provide the identical result

The p -value Approach to the Hypothesis Test

Specify what is meant by “likely” and “unlikely”

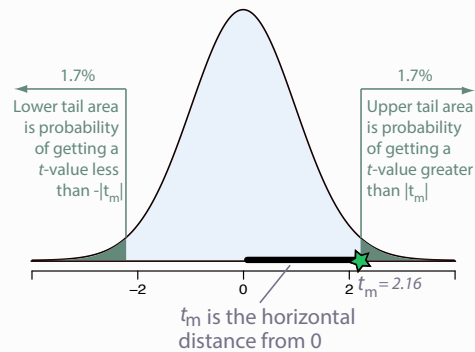
- ▶ Usually the definition of the threshold for “unlikely”, called alpha or α , is set at $\alpha = 0.05$
- ▶ Compare p -value to α to see if obtained t_m is unlikely
- ▶ The approach of comparing t_m to $t_{\alpha/2}$ is more “old school” now that obtaining the p -value is more feasible with computers

Statistical Decision

- ▶ p -value $> \alpha$, then fail to detect a difference from μ_0
- ▶ p -value $< \alpha$, then t_m is in the rejection region, so reject μ_0

Obtain p -value: Reject H_0

- ▷ Suppose $t_m = 2.16$
- ▷ t_m : Distance of m from μ_0 in terms of s_m
- ▷ For any one obtained $t_m > 0$, there is a corresponding upper-tail area, here 0.017
- ▷ To account for deviations in either direction, also consider corresponding lower-tail area
- ▷ Total area in both tails: $p\text{-value} = (2)(.017) = .034$

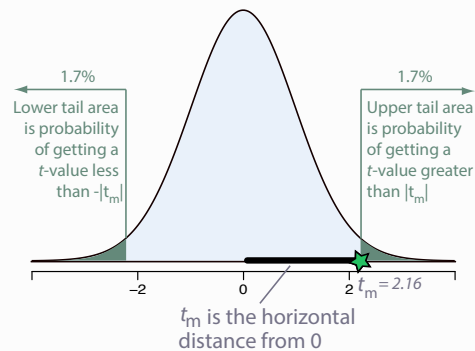


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Hypothesis Test: Compare Obtained p -value to α 11

Compare p -value with $\alpha = .05$: Reject H_0

- ▷ $p\text{-value} = .034 < \alpha = .05$
- IF μ_0 is true, THEN a m that far from μ_0 in either direction, or farther, has a low probability of occurring, only 0.034
- ▷ A low p -value means that, assuming μ_0 is true, the sample result is UNLIKELY
- ▷ So reject the null hypothesis and conclude that apparently $\mu \neq \mu_0$

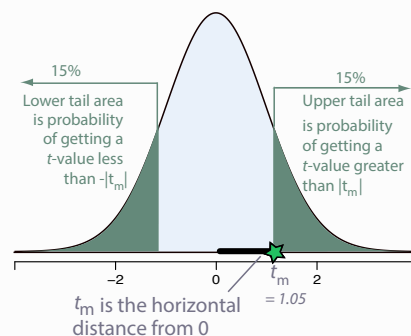


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Hypothesis Test: Compare Obtained p -value to α 12

Obtain p -value: Do not Reject H_0

- ▷ Suppose $t_m = 1.05$
- ▷ t_m : Distance of m from μ_0 in terms of s_m
- ▷ For any one $t_m > 0$, there is a corresponding upper-tail area, here 0.15
- ▷ To account for deviations in either direction, also consider corresponding lower-tail area
- ▷ Total area in both tails: $p\text{-value} = (2)(.15) = .30$



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Hypothesis Test: Compare Obtained p -value to α 13

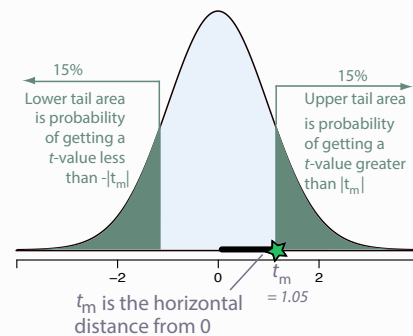
Compare p -value with $\alpha = .05$: Do *not* Reject H_0

▷ $p\text{-value} = .30 > \alpha = .05$

IF μ_0 is true, THEN a m that far from μ_0 , or farther, in either direction has a **reasonable probability** of occurring, 0.30

▷ A **high p -value** means that the **sample result is CONSISTENT** with the assumption of μ_0

▷ So do *not* reject the null hypothesis and conclude that **no difference** detected from μ_0



Interpretation of the Hypothesis Test

The statistical decision: Reject or Do not reject

- ▶ **Fail to Reject** the Null when ...
 - $p\text{-value} > \alpha$: Sample mean is close to μ_0
 - **Conclude**: No difference from μ_0 detected
 - Conclusion of no difference is NOT that μ_0 is true
- ▶ **Reject** the Null when ...
 - $p\text{-value} < \alpha$: Sample mean far from μ_0 , in rejection region
 - **Conclude**: A difference from the specified null hypothesize value of the mean is detected
 - For purposes of making a decision, simply detecting a difference does not provide any actionable information
 - Instead, what is important is the direction and extent of the difference, best evaluated with a confidence interval to follow-up the hypothesis test

p -value is a Conditional Probability

Distinguish between what we want and what we have

- ▶ **Conditional probability**: The probability of an event only when some other condition is true
- ▶ Notation: $P(A | B)$, read as the probability of event A given condition B
- ▶ A p -value is a conditional probability: $p\text{-value} = P(\text{data} | H_0)$
- ▶ The p -value is the probability of obtaining the data, and corresponding sample mean, *given* that the value of the mean specified by the null hypothesis is true
- ▶ Unfortunately, we actually want to know the probability that the null hypothesis is true given the data, $P(H_0 | \text{data})$
- ▶ ☹️ What we want is not what we get
- ▶ And knowing one conditional probability does not inform us as to the value of the other: $P(\text{data} | H_0) \neq P(H_0 | \text{data})$

Hypothesis Tests Never Provide the Probability of the Null

Conclusion of a hypothesis test is only qualitative

- ▶ **Key Concept:** Despite the precise calculation of a p -value, an hypothesis test provides a *qualitative result*, reasonable or unreasonable
- ▶ IF μ_0 is assumed true,
 - THEN when the sample result is unlikely, $p\text{-value} < .05$, conclude μ_0 is *unreasonable*
 - THEN when the sample result is likely, $p\text{-value} > .05$, conclude μ_0 is *reasonable*
- ▶ Computers can precisely calculate the p -value to as many decimal digits as desired
- ▶ The problem is that the probability of what we really want, the probability that μ_0 is the true value of μ , is not known, only qualitatively described as reasonable or unreasonable

R: One-sample Hypothesis Test

Analysis provided by `lessR` function `ttest`, or `tt`

- ▶ Analysis of variable Y from data: `> ttest(Y, mu=value)`
- ▶ **Ex:** To test that the mean weight of cereal in the cereal boxes is the specified 350g for the variable `Weight`
`> ttest(Weight, mu=350)`
- ▶ The output of `ttest` includes the relevant p -value
- ▶ The brief version, `tt_brief()`, provides just the descriptive statistics, the hypothesis test and confidence interval
- ▶ The output provides a graph of the smoothed histogram (densities) of the variable of interest
- ▶ Analysis of Y from sample size, mean and standard deviation:
`> ttest(n=value, m=value, s=value, mu=value)`

5.2c Application

Assess Average Weight of Boxed Cereal

Claim: Advertised weight in each box is 350 g

- ▶ Inherent process variability ensures that the contents of each box is typically more or less than 350 g
- ▶ But is there a systematic over- or under-fill as the cereal boxes are filled?
- ▶ Decision: On the production line, do we adjust the dial – up or down – to increase or decrease the average amount of cereal placed in each cereal box?

-
- ▶ Null Hypothesis is $H_0 : \mu = 350\text{g}$
 - ▶ Alternative Hypothesis is $H_1 : \mu \neq 350\text{g}$
-

- ▶ Data: Randomly sample 25 cereal boxes and measure the weight of the cereal in each box

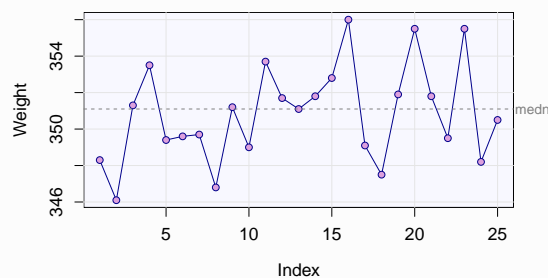
<http://lessRstats.com/data/boxweight.csv>

Weight
348.3
346.1
351.3
353.5
349.4
349.6
349.7
346.8
351.2
349.0
353.7
351.7
351.1
351.8
352.8
356.0
349.1
347.5
351.9
355.5
351.8
349.5
355.5
348.2
350.5

Stable Process Assumption of Y

Only meaningful to estimate a stable μ

- ▶ Are all data values of Y, Weight, from the same process?
- ▶ To answer, plot the run chart: `> LineChart(Weight)`

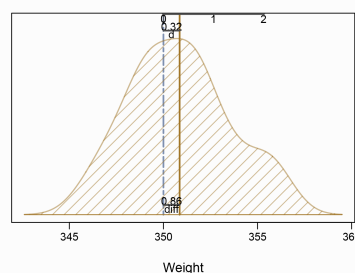


- ▶ Output appears to be random variation, so process stable
- ▶ Conclude: There is a stable μ to estimate

Normality Assumption of Sample Mean m

t -distribution probabilities only valid for normal m

- ▶ The distribution over repeated samples of m for n close to 30 is approximately normal unless Y is skewed
- ▶ To verify, examine the graph from the `ttest` function, a smoothed graph (the densities) of the distribution of Y



- ▶ The sample data are approximately normal so m , the mean of Weight, appears normal across multiple, hypothetical samples

Excel Template: Descriptive Statistics

	Description	Name	Value	Formula
INPUT: DESCRIPTIVE STATISTICS	count of data	n	25	COUNT(data)
	mean of data	mean	350.86	AVERAGE(data)
	standard dev of data	stdev	2.66	STDEV(data)

Traditional Notation:

$$n = 25$$

$$m = 350.86$$

$$s = 2.66$$

Same three basic descriptive statistics for calculating the confidence interval are also the basis for the hypothesis test

Basic Question:

Is sample mean of 350.860 g **close** to hypothesized mean of 350 g?

Excel Template: Standard Error

	Description	Name	Value	Formula
INPUT: DESCRIPTIVE STATISTICS	count of data	n	25	COUNT(data)
	mean of data	mean	350.860	AVERAGE(data)
	standard dev of data	stdev	2.660	STDEV(data)
	std error of mean	sterr	0.532	stdev/SQRT(n)

Traditional Notation:

$$s_m = \frac{s}{\sqrt{n}} = \frac{2.66}{\sqrt{25}} = 0.532$$

Same estimated standard error for calculating the confidence interval is also the basis for the hypothesis test

How many estimated standard errors separate 350.860 g from hypothesized mean of 350 g?

Excel Template: t-value

	Description	Name	Value	Formula
INPUT: DESCRIPTIVE STATISTICS	count of data	n	25	COUNT(data)
	mean of data	mean	350.860	AVERAGE(data)
	standard dev of data	stdev	2.660	STDEV(data)
	std error of mean	sterr	0.532	stdev/SQRT(n)
HYPOTHESIS TEST	hypothesized value	mu0	350	
	difference from null	diff	.86	mean-mu0
	t-value (distance)	t	1.617	diff/sterr
	p-value	p-value	0.119	TDIST(ABS(t),n-1,2)

Now specify the value of the hypothesized value, μ_0 , and then obtain the obtained t -value, followed by its corresponding p -value

$$t_m = \frac{m - \mu_0}{s_m} = \frac{350.86 - 350}{0.532} = \frac{0.860}{0.532} = 1.616$$

which yields a p -value = 0.119

R: Inference of the Mean, HT and CI, with lessR ttest

```
> tt_brief(Weight, mu=350)
```

```
Weight: n = 25, mean = 350.86, sd = 2.66
```

```
Hypothesized Value H0: mu = 350
```

```
Hypothesis Test of Mean:
```

```
t-value = 1.62, df = 24, p-value = 0.119
```

```
95% Confidence Interval for Mean: 349.76 to 351.96
```

Or, run the analysis directly from the summary statistics

```
> tt_brief(n=25, m=350.86, s=2.66, mu=350)
```

To obtain more information, run the full version of ttest

```
> ttest(Weight, mu=350)
```

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Result

Result of the Analysis

- ▶ **Statistical Result:** $t_m = \frac{m - \mu_0}{s_m} = 1.616$
 - The sample mean $m = 350.86\text{g}$ is **1.616 estimated standard errors above the hypothesized mean μ_0** which is $\mu = 350\text{g}$
 - Deviations from the hypothesized value of 350g in either direction are of interest, so is the distance of 1.616 standard errors or more from $\mu = 350\text{g}$ in *either* direction likely?
- ▶ **Statistical Result:** At $t_m = 1.616$, $df = 24$ the **$p\text{-value} \approx 12\%$**
 - For a true mean of $\mu = 350\text{g}$, **the probability of obtaining a value of m at or more than 0.86 grams on either side of 350g is 12%**
 - 12% does **not** indicate a particularly **rare event**

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Conclusion

Implications of the statistical decision

- ▶ **Statistical Decision:**
 $p\text{-value} = 0.119 > \alpha = .05$
so **do not reject the null hypothesized value of 350 g**,
the sample mean of 350.86 is close to 350
- ▶ **Interpretation:** **No difference** in the average weight of the contents of the cereal boxes **detected** from 350 g
- ▶ **Caveat:** **Not rejecting the null hypothesis** of $\mu = 350\text{ g}$ **does not imply** the true mean is 350 g
- ▶ **Managerial Decision:** **Do not adjust the dial** that controls how much cereal on average is placed in each cereal box
- ▶ **Follow-up Analysis:** When the **null hypothesis is not rejected**, something called **power analysis** should be conducted, explained later

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Appendix

Obtain the p -value Directly

Obtain the p -value

Technology moves on

- ▶ In the “old” days, accurate p -values were hard to obtain because a table of t -values does not typically provide enough probabilities for each t -distribution corresponding to a specific degrees of freedom
- ▶ Today p -values are automatically provided by the computer when conducting an hypothesis test, with applications such as R and Excel
- ▶ The p -value section of computer output of the hypothesized test is usually considered the most crucial part of the displayed analysis
- ▶ Also can obtain a p -value with programs such as R and Excel apart from a specific data analysis, analogous to the “old school” method of looking up a value in a printed t -table, shown in the Appendix

R: Manually Obtain a p -value

The R `pt` function

- ▶ Use the probability t -distribution function, `pt`, which by default provides the cumulative probability, the “lower tail” of the distribution, appropriate for obtained negative values of t_m
- ▶ For the usual test that investigates possibilities in both tails, multiply the obtained tail probability by two to get the probability of a deviation as large or larger in either tail
- ▶ Ex: Obtained $t_m = -1.616$, $df=24$

```
> 2*pt(-1.616, df=24) returns 0.119
```
- ▶ Set `lower.tail=FALSE` to get the upper tail probability, which is usually desired for positive values of t_m
- ▶ Ex: Obtained $t_m = 1.616$, $df=24$

```
> 2*pt(1.616, df=24, lower.tail=FALSE) returns 0.119
```

Excel: Manually Obtain a p -value

Use the TDIST function

- ▶ Use the **TDIST**ribution function to get the probability of one or both tails of the specified t -distribution

```
=TDIST(ABS(t-value), df, number of tails)
```

- ▶ Only positive t -statistics can be specified, so first call the ABSsolute value function or only enter a positive t -value
- ▶ Ex: for $t_m = 1.616$, $df = 24$

```
=TDIST(ABS(1.616), 24, 2) returns 0.119
```

- ▶ Excel has no one-sample t -test function
- ▶ An Excel template for this t -test which uses the **TDIST** function is provided in the following application

Index Subtract 2 from each listed value to get the Slide

For the computer analysis use the
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▶ The End